

2021-08

Type-1 OWA Operators in Aggregating Multiple Sources of Uncertain Information: Properties and Real-World Applications in Integrated Diagnosis

Zhou, Shang-Ming

<http://hdl.handle.net/10026.1/18227>

10.1109/TFUZZ.2020.2992909

IEEE Transactions on Fuzzy Systems

Institute of Electrical and Electronics Engineers (IEEE)

All content in PEARL is protected by copyright law. Author manuscripts are made available in accordance with publisher policies. Please cite only the published version using the details provided on the item record or document. In the absence of an open licence (e.g. Creative Commons), permissions for further reuse of content should be sought from the publisher or author.

Supplemental Material for the Manuscript - “Type-1 OWA Operators in Aggregating Multiple Sources of Uncertain Information : Properties and Real World Applications in Integrated Diagnosis”

Shang-Ming Zhou, Member, IEEE, Francisco Chiclana, Robert I. John, Senior Member, IEEE, Jon M. Garibaldi, Senior Member, IEEE, and Lin Huo

I. MORE EXAMPLES OF TYPE-1 OWA AGGREGATIONS

Example 1 (A T1OWA Operator). *Figure 1 depicts the results of a T1OWA operator to aggregate four fuzzy sets. Figure 1a shows the linguistic weights. Figure 1b shows the aggregated fuzzy set objects and aggregation result.*

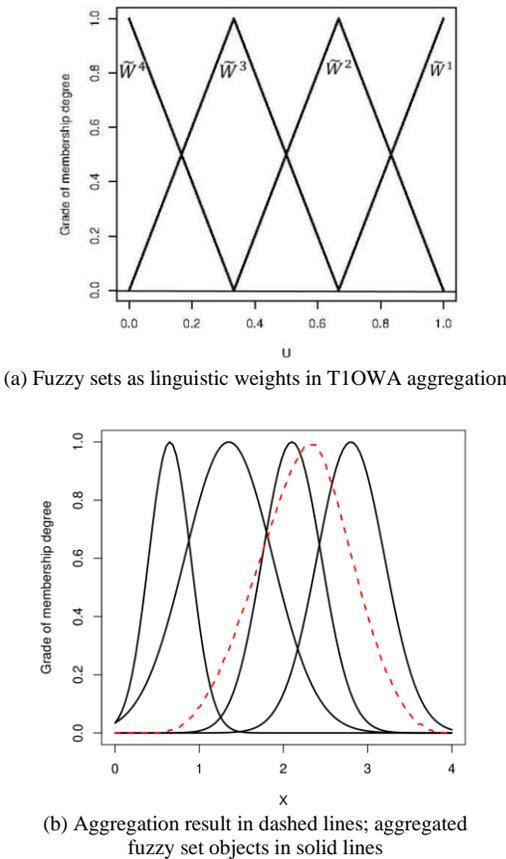


Fig. 1: A T1OWA operator

Example 2 (Join Operator). *Figure 2 illustrates the result of join operator to aggregate fuzzy sets as the extended maximum of fuzzy sets, a special T1OWA operator.*

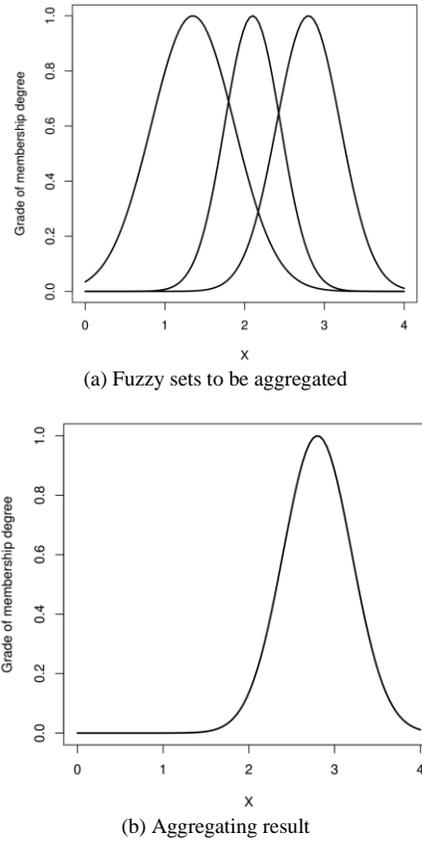


Fig. 2: Aggregation of T1OWA operator as join

Example 3 (Join-Like T1OWA Operator). *The join-like T1OWA operation requests that the first linguistic weight move towards the singleton fuzzy set, $\tilde{1}$, and all others towards the singleton fuzzy set $\tilde{0}$ (Figure 4a and Figure 4b). Using these weights to aggregate the fuzzy sets, Figure 3c illustrates how this T1OWA exhibits the join-like behaviour when aggregating three fuzzy sets. Clearly, the final aggregation fuzzy set is closer to the rightmost fuzzy set (i.e., the fuzzified maximum)*

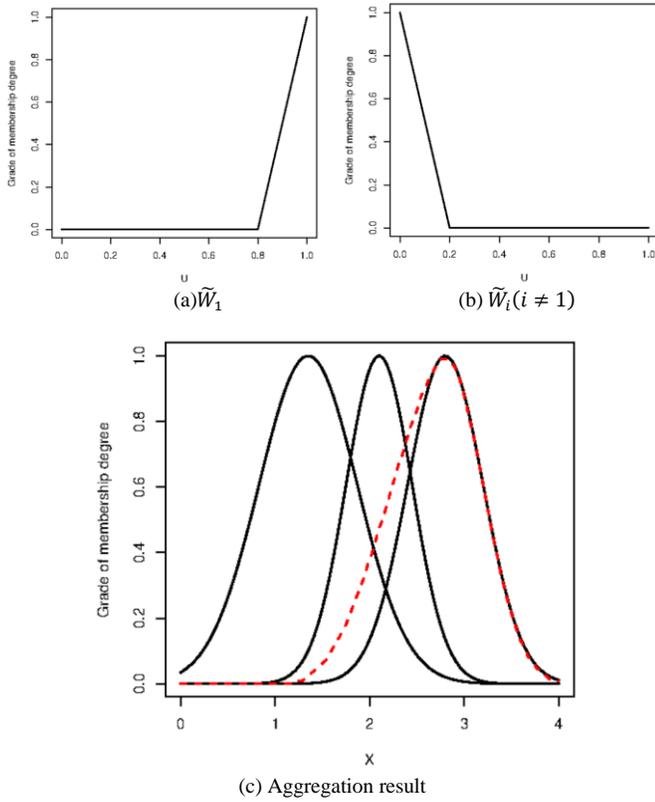


Fig. 3: A join-like T1OWA operator. (a)(b)-Linguistic weights;(c)-Aggregation result. Solid lines representing the aggregated objects; dashed line representing the aggregation result.

Example 4 (Joinness of Join-Like T1OWA Operator). The Figure 4c shows an example of the joinness of the join-like T1OWA operator defined by the fuzzy weights in Figure 4a and Figure 4b.

Example 5 (Meet Operator). Figure 5 illustrates the result of meet operator to aggregate fuzzy sets as the extended minimum of fuzzy sets, a special T1OWA operator.

Example 6 (Meet-Like T1OWA Operator). Meet-like T1OWA operation requests that the last linguistic weight to be close to $\bar{1}$ (Figure 4a), and all other weights close to $\bar{0}$ (Figure 4b). Figure 6 illustrates how this T1OWA exhibits the meet-like behaviour when aggregating three fuzzy sets. Clearly, the final aggregation fuzzy set is closer to the leftmost fuzzy set (i.e., the fuzzified minimum).

Example 7 (Counterexample of α -Equivalently Ordered Relation). Figure 7 shows the groups of fuzzy numbers $\{\tilde{B}^1, \tilde{B}^2, \tilde{B}^3\}$ are not α -equivalently ordered with another group $\{\tilde{A}^1, \tilde{A}^2, \tilde{A}^3\}$: Figure 7a shows that at the $\alpha = 0.2$ level: $\tilde{A}_{0.2+}^3 \geq \tilde{A}_{0.2+}^2 \geq \tilde{A}_{0.2+}^1$, i.e., the permutation operator $\sigma = (3; 2; 1)$, but $\tilde{B}_{0.2+}^3 \geq \tilde{B}_{0.2+}^1 \geq \tilde{B}_{0.2+}^2$; while Fig.7b shows that $\tilde{A}_{0.2-}^3 \geq \tilde{A}_{0.2-}^2 \geq \tilde{A}_{0.2-}^1$, i.e., the permutation operator $\eta = (3; 2; 1)$, but $\tilde{B}_{0.2-}^3 \geq \tilde{B}_{0.2-}^1 \geq \tilde{B}_{0.2-}^2$.

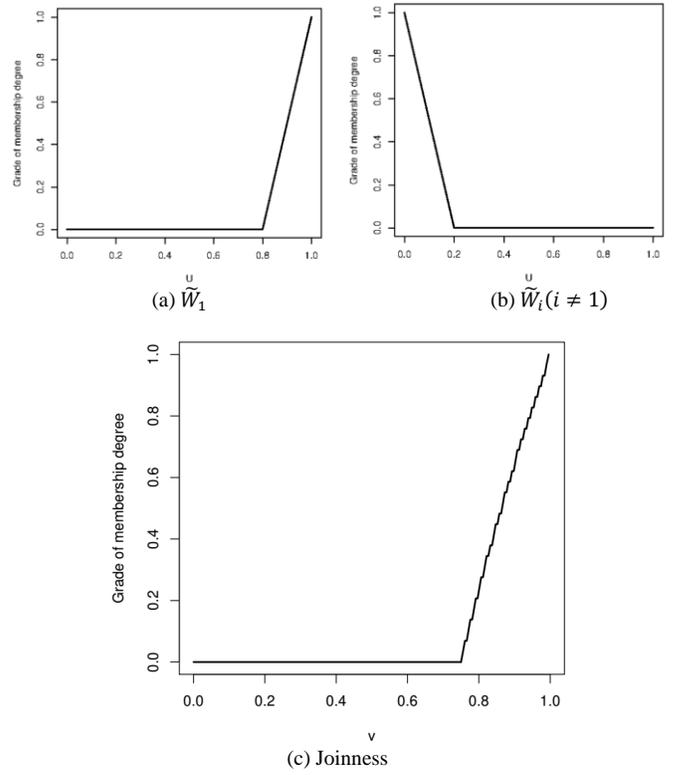


Fig. 4: Joinness of a join-like T1OWA operator

Example 8. Supposing the numerical domains $U = \{0.0, 0.5, 1.0\}$ and $X = \{0.0, 1.0, 2.0\}$. Let the given linguistic

weights $\tilde{W} = \left(\mu_{\tilde{W}}(\omega_i) \right)_{\omega_i \in U}$ on U be

$$\tilde{W}^1 = \begin{pmatrix} 0.0 & 0.5 & 1.0 \\ 1.0 & 0.5 & 0.0 \end{pmatrix}; \tilde{W}^2 = \begin{pmatrix} 0.0 & 0.5 & 1.0 \\ 0.0 & 1.0 & 0.0 \end{pmatrix};$$

$$\tilde{W}^3 = \begin{pmatrix} 0.0 & 0.5 & 1.0 \\ 0.0 & 0.5 & 1.0 \end{pmatrix}$$

and the fuzzy sets to be aggregated be

$$\tilde{A}^1 = \begin{pmatrix} 0.0 & 1.0 & 2.0 \\ 0.0 & 0.5 & 1.0 \end{pmatrix}; \tilde{A}^2 = \begin{pmatrix} 0.0 & 1.0 & 2.0 \\ 1.0 & 0.5 & 0.0 \end{pmatrix};$$

$$\tilde{A}^3 = \begin{pmatrix} 0.0 & 1.0 & 2.0 \\ 0.0 & 1.0 & 0.0 \end{pmatrix}.$$

Let G be the aggregation result of this T1OWA defined by the above linguistic weights. The aggregation result by the join operator J can be calculated as follows.

Case I. $\alpha = 0.0$

As demonstrated by the Example 1 in [5], one can calculate $G_\alpha = \{0.0, 1.0, 2.0\}$. But

$$J_\alpha(\tilde{A}_\alpha^1, \tilde{A}_\alpha^2, \tilde{A}_\alpha^3)_+ = \max(\tilde{A}_{\alpha+}^1, \tilde{A}_{\alpha+}^2, \tilde{A}_{\alpha+}^3) = 2.0$$

$$J_\alpha(\tilde{A}_\alpha^1, \tilde{A}_\alpha^2, \tilde{A}_\alpha^3)_- = \max(\tilde{A}_{\alpha-}^1, \tilde{A}_{\alpha-}^2, \tilde{A}_{\alpha-}^3)$$

$$=0.0$$

So $G_{\alpha-} \leq J_{\alpha-}$, $G_{\alpha+} \leq J_{\alpha+}$.

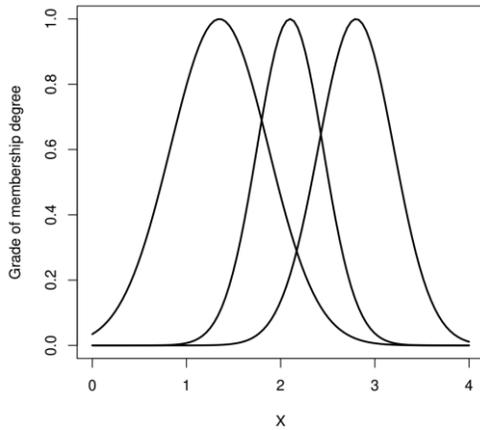
Case II. $\alpha = 0.5$

As demonstrated by the Example 1 in [5], one can calculate $G_{\alpha} = \{1.0\}$. But

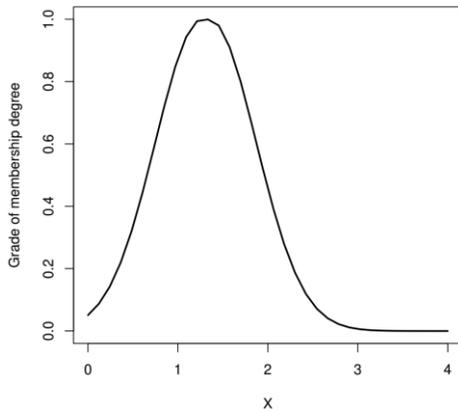
$$\begin{aligned} J_{\alpha}(\tilde{A}_{\alpha}^1, \tilde{A}_{\alpha}^2, \tilde{A}_{\alpha}^3)_{+} &= \max(\tilde{A}_{\alpha+}^1, \tilde{A}_{\alpha+}^2, \tilde{A}_{\alpha+}^3) \\ &= \max(2.0, 1.0, 1.0) \\ &= 2.0 \end{aligned}$$

$$\begin{aligned} J_{\alpha}(\tilde{A}_{\alpha}^1, \tilde{A}_{\alpha}^2, \tilde{A}_{\alpha}^3)_{-} &= \max(\tilde{A}_{\alpha-}^1, \tilde{A}_{\alpha-}^2, \tilde{A}_{\alpha-}^3) \\ &= \max(1.0, 0.0, 1.0) \\ &= 1.0 \end{aligned}$$

So $G_{\alpha-} \leq J_{\alpha-}$, $G_{\alpha+} \leq J_{\alpha+}$.



(a) Fuzzy sets to be aggregated



(b) Aggregating result

Fig. 5: Aggregation of TIOWA operator as meet

Case III. $\alpha = 1.0$

As demonstrated by the Example 1 in [5], one can calculate $G_{\alpha} = \emptyset$. But

$$\begin{aligned} J_{\alpha}(\tilde{A}_{\alpha}^1, \tilde{A}_{\alpha}^2, \tilde{A}_{\alpha}^3)_{+} &= \max(\tilde{A}_{\alpha+}^1, \tilde{A}_{\alpha+}^2, \tilde{A}_{\alpha+}^3) \\ &= \max(2.0, 0.0, 1.0) \\ &= 2.0 \end{aligned}$$

$$\begin{aligned} J_{\alpha}(\tilde{A}_{\alpha}^1, \tilde{A}_{\alpha}^2, \tilde{A}_{\alpha}^3)_{-} &= \max(\tilde{A}_{\alpha-}^1, \tilde{A}_{\alpha-}^2, \tilde{A}_{\alpha-}^3) \\ &= \max(2.0, 0.0, 1.0) \\ &= 2.0 \end{aligned}$$

So $G_{\alpha-} \leq J_{\alpha-}$, $G_{\alpha+} \leq J_{\alpha+}$.

Then according to the Definition about partial order relation of fuzzy sets, we have $J \succcurlyeq G$. Similarly, $G \succcurlyeq M$, the meet operator.

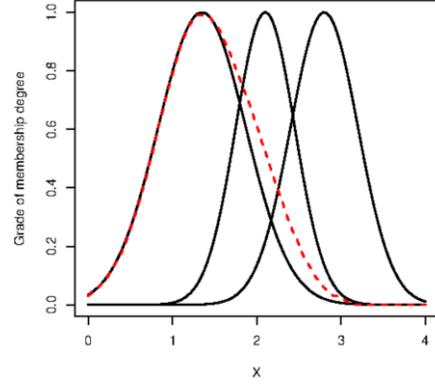
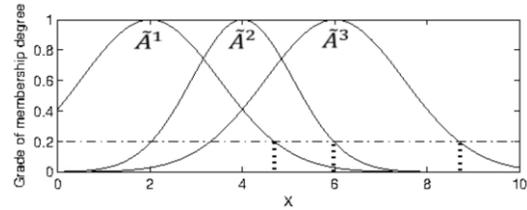
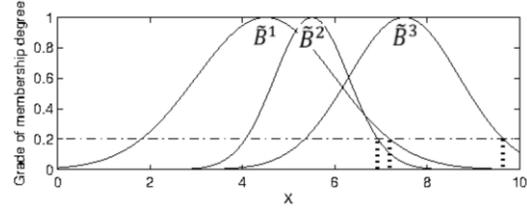


Fig. 6: Aggregation result by a meet-like TIOWA operator. Solid lines representing the aggregated objects; dashed line representing the aggregation result.



$$(a) \tilde{A}_{0.2+}^3 \geq \tilde{A}_{0.2+}^2 \geq \tilde{A}_{0.2+}^1 \text{ and } \tilde{B}_{0.2+}^3 \geq \tilde{B}_{0.2+}^1 \geq \tilde{B}_{0.2+}^2$$



$$(b) \tilde{A}_{0.2-}^3 \geq \tilde{A}_{0.2-}^2 \geq \tilde{A}_{0.2-}^1 \text{ and } \tilde{B}_{0.2-}^3 \geq \tilde{B}_{0.2-}^1 \geq \tilde{B}_{0.2-}^2$$

Fig. 7: Fuzzy numbers $\tilde{B}^1, \tilde{B}^2, \tilde{B}^3$ (bottom) not α -equivalently ordered with $\tilde{A}^1, \tilde{A}^2, \tilde{A}^3$ (up) separately

II. CASE STUDY

A. The TIOWA based non-stationary fuzzy system

Figure 8 illustrates the structure of TIOWA based nonstationary fuzzy inference system, the TIOWA operator is used to aggregate the multiple fuzzy decisions from non-stationary fuzzy inference engine.

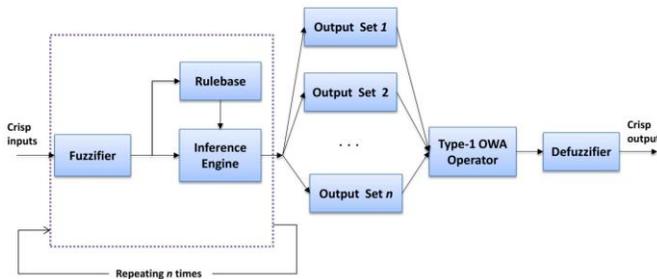


Fig. 8: TIOWA based non-stationary fuzzy inference system.

B. Fuzzy sets of the variables - "plaGlu", "BMI" and "Outcome"

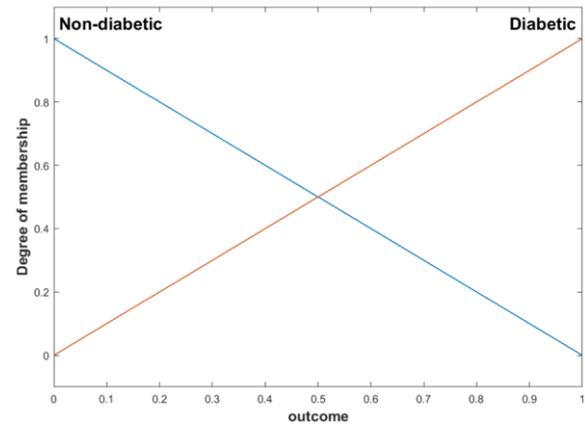
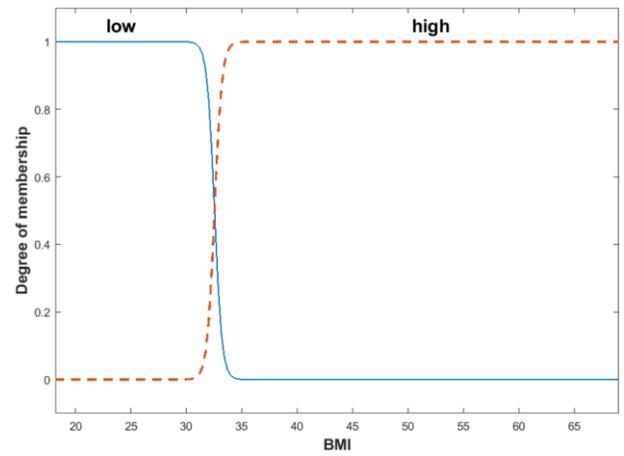
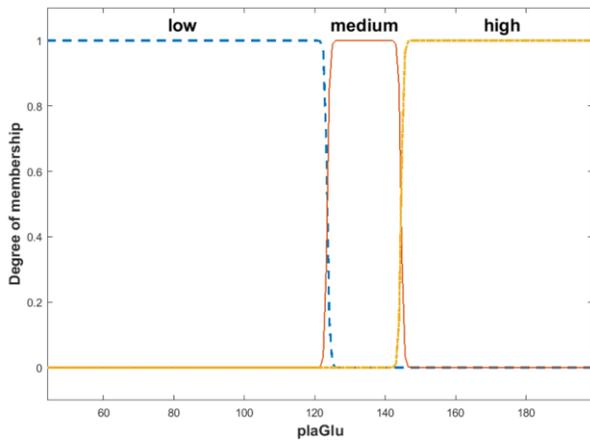


Fig. 9: Membership functions of fuzzy sets for the attributes of *plaGlu*, *BMI* and *outcome*.

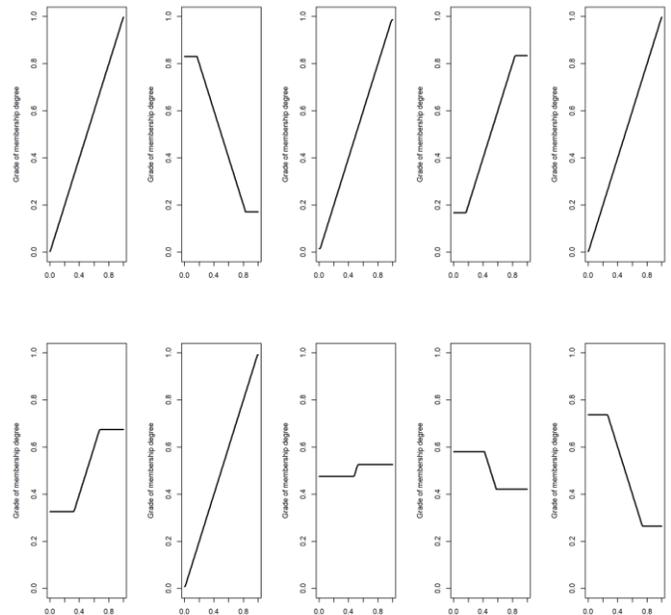


Fig. 10: Example of 10 fuzzy output decisions by the nonstationary fuzzy system for a patient.

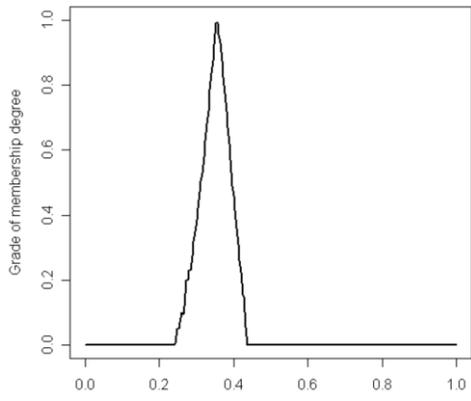


Fig. 11: Joinness of the type-2 quantifier “*most*” guided T1OWA operator.